

# Foundations of Artificial Intelligence

## 6. State-Space Search: Representation of State Spaces

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6.1 Representation of State Spaces

6.2 Explicit Graphs

6.3 Declarative Representations

6.4 Black Box

6.5 Summary

## State-Space Search: Overview

Chapter overview: state-space search

- ▶ 5.–7. Foundations
  - ▶ 5. State Spaces
  - ▶ 6. Representation of State Spaces
  - ▶ 7. Examples of State Spaces
- ▶ 8.–12. Basic Algorithms
- ▶ 13.–19. Heuristic Algorithms

## 6.1 Representation of State Spaces

## Representation of State Spaces

- ▶ practically interesting state spaces are often **huge** ( $10^{10}$ ,  $10^{20}$ ,  $10^{100}$  states)
- ▶ How do we **represent** them, so that we can efficiently deal with them algorithmically?

three main options:

- 1 as **explicit** (directed) graphs
- 2 with **declarative** representations
- 3 as a **black box**

German: explizite Graphen, deklarative Repräsentationen, Black Box

## 6.2 Explicit Graphs

## State Spaces as Explicit Graphs

### State Spaces as Explicit Graphs

represent state spaces as **explicit directed graphs**:

- ▶ vertices = states
- ▶ directed arcs = transitions

↔ represented as **adjacency list** or **adjacency matrix**

German: Adjazenliste, Adjazenzmatrix

Example (explicit graph for 8-puzzle)

puzzle8.graph

## State Spaces as Explicit Graphs: Discussion

discussion:

- ▶ **impossible** for **large** state spaces (too much space required)
- ▶ if spaces small enough for explicit representations, solutions easy to compute: **Dijkstra's algorithm**  
 $O(|S| \log |S| + |T|)$
- ▶ interesting for time-critical **all-pairs-shortest-path** queries  
(examples: route planning, path planning in video games)

## 6.3 Declarative Representations

## State Spaces with Declarative Representations

### State Spaces with Declarative Representations

represent state spaces **declaratively**:

- ▶ **compact** description of state space as input to algorithms  
 ↪ state spaces **exponentially larger** than the input
- ▶ algorithms directly operate on compact description
- ↪ allows automatic reasoning about problem:  
 reformulation, simplification, abstraction, etc.

**Example (declarative representation for 8-puzzle)**

puzzle8-domain.pddl + puzzle8-problem.pddl

## 6.4 Black Box

## State Spaces as Black Boxes

### State Spaces as Black Boxes

Define an **abstract interface** for state spaces.

For state space  $\mathcal{S} = \langle S, A, cost, T, s_0, S_* \rangle$

we need these methods:

- ▶ **init()**: generate initial state  
 result: state  $s_0$
- ▶ **is\_goal(s)**: test if  $s$  is a goal state  
 result: **true** if  $s \in S_*$ ; **false** otherwise
- ▶ **succ(s)**: generate applicable actions and successors of  $s$   
 result: sequence of pairs  $\langle a, s' \rangle$  with  $s \xrightarrow{a} s'$
- ▶ **cost(a)**: gives cost of action  $a$   
 result:  $cost(a) (\in \mathbb{N}_0)$

**Remark:** we will extend the interface later  
 in a small but important way

## State Spaces as Black Boxes: Example and Discussion

### Example (Black Box Representation for 8-Puzzle)

demo: `puzzle8.py`

- ▶ in the following: focus on black box model
- ▶ explicit graphs only as illustrating examples
- ▶ near end of semester: declarative state spaces (classical planning)

## 6.5 Summary

## Summary

- ▶ state spaces often huge ( $> 10^{10}$  states)  
↪ how to represent?
- ▶ explicit graphs: adjacency lists or matrices;  
only suitable for small problems
- ▶ declaratively: compact description as input  
to search algorithms
- ▶ black box: implement an abstract interface